



Matching Daily Data on Resource Use to Fire Suppression Costs

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ABSTRACT

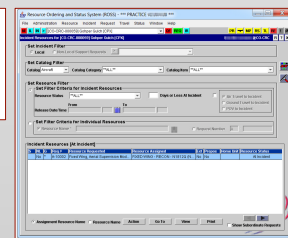
- The quality of an analysis is critically dependent on the quality of the underlying data.
- Measuring cost-effectiveness of fuel treatments on Federal lands remains an important research and policy question.
- Much of past literature focuses on fire-level analysis but using daily data provides an improved analysis lens for identifying fuel treatment effects on suppression costs.
- A major hurdle to daily analyses has been accessing daily suppression cost data.
- For 62 fires (1,125 fire-days), daily suppression cost data from ISuite application were successfully matched to resource use data in the National Interagency Resource Ordering and Status System (ROSS).
- While accessing ISuite data is difficult, ROSS data are somewhat easier to access.
- Based on our matched sample of data we find that a simple count of ROSS category lines for each day, explains a high percentage of variation in daily suppression costs.
- The ISuite cost data, therefore, seem to be capturing resource use well.

THE PROBLEM

- Researchers analyzing cost-effectiveness of fuel treatments have consistently been faced with limited availability of adequate suppression cost data.
- As Thompson and Anderson (2015, p. 166) put it: "knowledge gaps and data limitations have precluded direct quantification of the influence of past fuel treatment investments on wildfire suppression expenditures."
- Daily costs are needed to identify fuel treatment effects on costs as the fire moves across the landscape.
- ISuite application collects an array of daily suppression cost data but the Forest Service does not have sufficient resources to make these data available for researchers.
- The ROSS data on suppression resource use is more readily available and has a benefit in that units of resources are equivalent across states.
- Nevertheless, for economic science using monetary values facilitates analysis by bringing resource use into one measurable dimension.

Matched Fire-Days Data by State for 2008-2012

State	Fires	Days	Total Cost (\$1,000s)	Cost/Fire (\$1,000s)	Cost/Day (\$1,000s)
AZ	3	22	\$3,779	\$1,260	\$171.78
CA	10	253	\$271,555	\$27,503	\$1,073.34
CO	2	34	\$48,837	\$24,419	\$1,436.39
ID	3	21	\$6,742	\$2,247	\$321.04
MN	2	73	\$24,022	\$12,011	\$329.07
MT	7	105	\$26,965	\$3,852	\$256.81
NC	1	46	\$12,110	\$12,110	\$263.27
NM	1	14	\$4,227	\$4,227	\$301.92
NV	7	75	\$18,433	\$2,633	\$245.78
OK	1	6	\$586	\$586	\$97.60
OR	8	160	\$75,255	\$9,407	\$470.34
SD	1	4	\$540	\$540	\$134.97
TX	1	4	\$1,749	\$1,749	\$437.27
UT	6	129	\$35,524	\$5,921	\$275.38
VA	2	98	\$19,349	\$9,674	\$197.43
WA	3	31	\$11,247	\$3,749	\$362.82
WY	4	50	\$20,926	\$5,232	\$418.53
Total	62	1,125	\$581,847		
Average	3.6	66.2	\$34,226	\$7,478	\$400



e-ISuite Cost

Unit	Resource	Cost	Unit	Resource	Cost
1	Airtanker Type 1	54.445	1	Airtanker Type 1	54.445
2	Airtanker Type 2	42.887	2	Airtanker Type 2	42.887
3	Airtanker Type 3	29.411	3	Airtanker Type 3	29.411
4	Airtanker Type 4	22.496	4	Airtanker Type 4	22.496
5	Airtanker Type 5	38.021	5	Airtanker Type 5	38.021
6	Airtanker Type 6	22.835	6	Airtanker Type 6	22.835
7	Airtanker Type 7	4.590	7	Airtanker Type 7	4.590
8	Airtanker Type 8	11.459	8	Airtanker Type 8	11.459
9	Airtanker Type 9	24.377	9	Airtanker Type 9	24.377
10	Airtanker Type 10	21.378	10	Airtanker Type 10	21.378
11	Airtanker Type 11	25.980	11	Airtanker Type 11	25.980
12	Airtanker Type 12	22.856	12	Airtanker Type 12	22.856
13	Airtanker Type 13	3.277	13	Airtanker Type 13	3.277
14	Airtanker Type 14	13.593	14	Airtanker Type 14	13.593
15	Airtanker Type 15	4.284	15	Airtanker Type 15	4.284
16	Airtanker Type 16	3.160	16	Airtanker Type 16	3.160
17	Airtanker Type 17	4.379	17	Airtanker Type 17	4.379
18	Airtanker Type 18	11.731	18	Airtanker Type 18	11.731
19	Airtanker Type 19	-9.221	19	Airtanker Type 19	-9.221
20	Airtanker Type 20	11.418	20	Airtanker Type 20	11.418
21	Airtanker Type 21	88.925	21	Airtanker Type 21	88.925
22	Airtanker Type 22	1.125	22	Airtanker Type 22	1.125
23	Airtanker Type 23	0.965	23	Airtanker Type 23	0.965
24	Airtanker Type 24	1.125	24	Airtanker Type 24	1.125
25	Airtanker Type 25	0.984	25	Airtanker Type 25	0.984
26	Airtanker Type 26	0.990	26	Airtanker Type 26	0.990
27	Airtanker Type 27	0.986	27	Airtanker Type 27	0.986

ANALYSIS

- Pooled ordinary least squares estimation is used as a baseline to explore how well the variation in ROSS *Categories* per fire-day explains the variation in ISuite daily suppression *Costs* for fire *f* on day *t*.

$$Costs_{f,t} = \beta_0 + \beta_1 * Categories_{f,t} + \epsilon_{f,t}$$

- Additional analysis included fire fixed effects, allowing the constant term β_0 to vary by fire.
- The Hausman test rejected the random effects model in favor of fixed effects in this equation.
- Over 500 possible ROSS *Categories* are available in the dataset but with 1,125 observations, not all these could conceivably be included in the analysis.
- Regression analysis used 53 *Categories* and reports coefficients for a subset of those:

VARIABLES	(1) Pooled OLS	(2) Fixed Effects	(3) Fixed Effects	(4) Fixed Effects
	All Fires	All Fires	Cal Fires	Non-Cal Fires
Airtanker Type1	54.445***	45.464***	56.139***	52.139***
Airtanker Type2	42.887***	42.730***	62.911***	51.619***
Airtanker Type3	29.411***	27.264***	33.162***	19.205**
Airtanker Type4	22.496***	24.760***	123.108*	18.785***
Airtanker Type5	-38.021***	-21.126**	-43.410	-16.429***
Airtanker Type6	22.835***	26.238***	30.472***	18.685***
Airtanker Type7	4.590**	5.312**	20.163**	4.764**
Airtanker Type8	11.459***	15.837***	15.416	11.514***
Airtanker Type9	24.377***	26.255***	34.160***	17.907
Airtanker Type10	21.378	26.257***	130.878***	
Airtanker Type11	25.980***	17.823***	15.209***	58.516**
Airtanker Type12	22.856**	27.013***	82.190	17.642***
Airtanker Type13	3.277	-12.228***	0.683	18.880***
Airtanker Type14	13.593	31.460***	53.564	-2.609
Airtanker Type15	4.284***	8.035***	8.940*	4.439**
Airtanker Type16	3.160***	2.757***	-6.129	3.074**
Airtanker Type17	4.379	5.510	-9.488	8.693***
Airtanker Type18	11.731***	4.893	-42.160***	28.420***
Airtanker Type19	-9.221*	-16.393***	-15.679	-2.783
Airtanker Type20	11.418**	11.428*	-25.423	15.766***
Airtanker Type21	88.925***	44.195***	109.962***	0.169
Observations	1,125	1,125	253	872
R-squared	0.965	0.984	0.990	0.986

- Type 1 resources, typically although not always, increase *Costs* more than Type 2 resources.
- Some ROSS categories might be capturing specific characteristics of the fire (e.g. Airtanker Type 4 Single Engines tend to be used on BLM lands and therefore, the negative coefficient might be capturing lower suppression costs of BLM fires).
- Estimates are different for California.

RESULTS

- Daily observations from ISuite match up well with ROSS data, albeit not perfectly.
- About 96% of variation in daily *Costs* (from ISuite) can be explained by variation in ROSS resource use *Categories*, for the matched days.
- Using more detailed (and a larger number of) ROSS *Categories* does not increase explanatory power much.
- The relationship between ROSS *Categories* and *Costs* varies across states (specifically California vs. not).
- In Fixed Effects model, the highest intercept terms are estimated for following fires: High Park – CO, North Pass – CA, Salmon – NV, and Sayre – CA.
- These are preliminary results and warrant further evaluation of fire-level data to ensure that data entry was comparable across fires.

CONCLUSIONS

- A large percentage of variation in ISuite daily *Costs* can be explained by variation in ROSS resource use *Categories*.
- As the number of ROSS *Categories* is changed from a high of several hundred to a low of three dozen, the fit of the model remains approximately the same.
- Including fire fixed effects, after ROSS *Categories*, increases explanatory power of the model only slightly.
- The coefficient estimates are somewhat sensitive to how aircraft use is measured (in hours per day vs. number of aircraft) but main results remain unchanged.
- Preliminary analysis confirms that ISuite suppression cost data for our sample of 62 fires match well with ROSS resource use.

Reference

Thompson, M.P.; Anderson, N.M. 2015. "Modeling fuel treatment impacts on fire suppression cost savings: A review," *California Agriculture*, 69(3), 164-170.

Acknowledgements

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